

Volume-of-Fluid Discretization Methods for PDE in Irregular Domains

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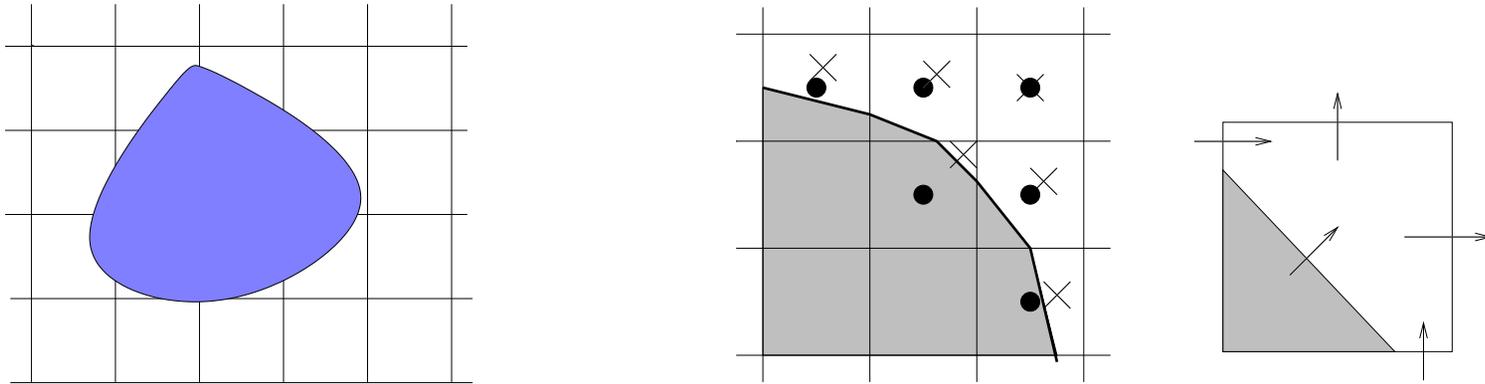
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Cartesian Grid Representation of Irregular Boundaries

Based on nodal-point representation (Shortley and Weller, 1938) or finite-volume representation (Noh, 1964).

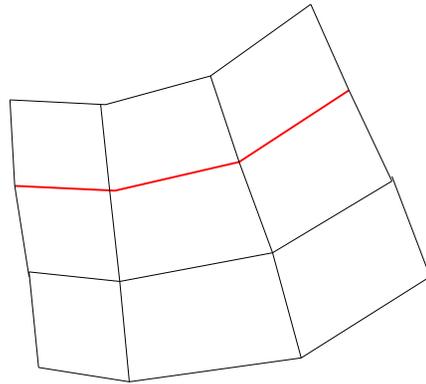


Advantages of underlying rectangular grid:

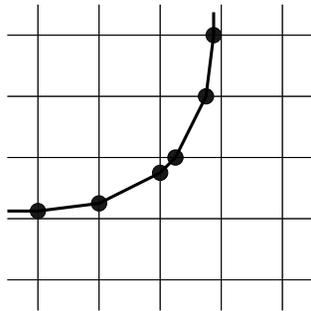
- Grid generation is tractable (Aftosmis, Berger, and Melton, 1998).
- Good discretization technology, e.g. well-understood consistency theory for finite differences, geometric multigrid for solvers.
- Straightforward coupling to structured AMR (Chern and Colella, 1987; Young et. al., 1990; Berger and Leveque, 1991).

Lagrangian vs. Eulerian Representations of Free Surfaces

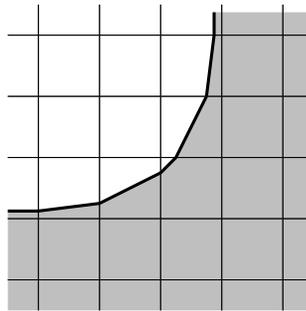
Lagrangian:



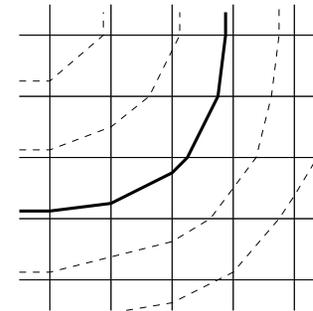
Eulerian:



Polygonal
(LANL, 1950s)



Volume of fluid
(LANL, LLNL, 1960s)

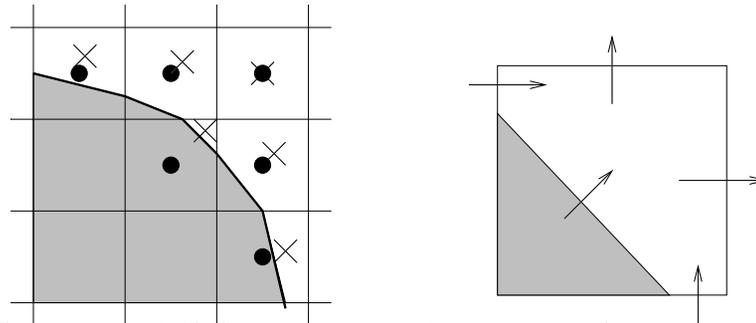


Level Set
(Osher & Sethian, 1988)

Finite-Volume Discretization - Fixed Boundaries

Consider PDEs written in conservation form:

$$\nabla \cdot (\nabla \phi) = \rho \qquad \frac{\partial U}{\partial t} + \nabla \cdot \vec{F}(U) = 0$$



- Primary dependent variables approximate values at centers of Cartesian cells. Extension of smooth functions to covered region exists, and extension operator is a bounded operator on the relevant function spaces.
- Divergence theorem over each control volume leads to “finite volume” approximation for $\nabla \cdot \vec{F}$:

$$\nabla \cdot \vec{F} \approx \frac{1}{\kappa h^d} \int \nabla \cdot \vec{F} dx = \frac{1}{\kappa h} \sum \alpha_s \vec{F}_s \cdot \vec{n}_s + \alpha_B \vec{F} \cdot \vec{n}_B \equiv D \cdot \vec{F}$$

- Away from the boundaries, method reduces to standard conservative finite difference discretization.

- If $\vec{F}_s \cdot \vec{n}_s$ approximates the value at the centroid to $O(h^2)$, then the truncation error $\tau = D \cdot \vec{F} - \nabla \cdot \vec{F}$ is given by

$$\tau = O(h^2) \text{ at interior cells (if approximation is smooth).}$$

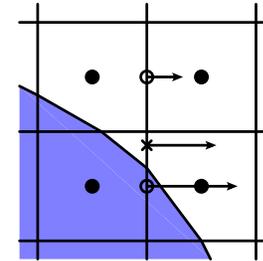
$$= O\left(\frac{h}{\kappa}\right) \text{ at irregular control volumes.}$$

Poisson's Equation

$$\Delta\phi = \rho \Rightarrow L^h\phi^h = \rho^h$$

$$L^h = D\vec{F}, \quad \vec{F} \approx \nabla\phi$$

\vec{F} computed using linear interpolation of centered difference approximations to derivatives of ϕ .



$$L^h(\phi^h)_i = \frac{1}{\kappa_i} \sum_{s \in \mathcal{S}_i} a_s \phi_s^h$$

$$a_s = O\left(\frac{1}{h^2}\right) \text{ uniformly w.r.t. } \kappa$$

The small denominator can be eliminated by diagonal scaling, eliminating the obvious potential conditioning problem: we solve

$$\kappa_i L^h(\phi^h)_i = \kappa_i \rho_i^h$$

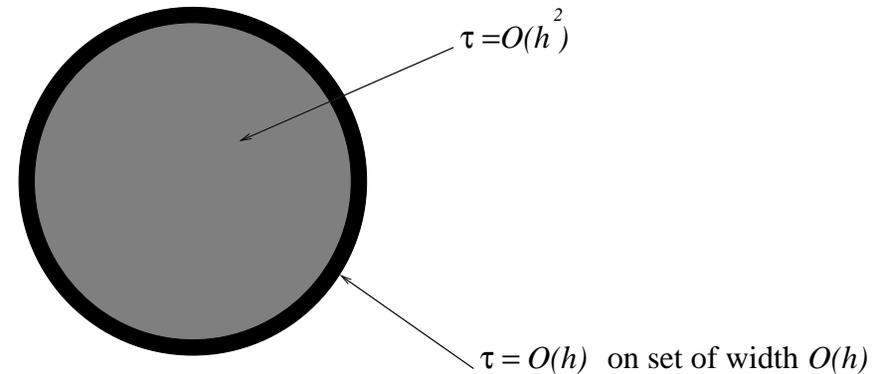
Modified Equation Analysis

Error equation: $\phi^{exact,h} = \phi^h + (L^h)^{-1}(\tau)$

Modified equation: $\epsilon = (L^h)^{-1}(\tau) \approx \Delta^{-1}\tilde{\tau}$

where $\tilde{\tau}$ is some extension of τ , e.g.

piecewise constant on each control volume.

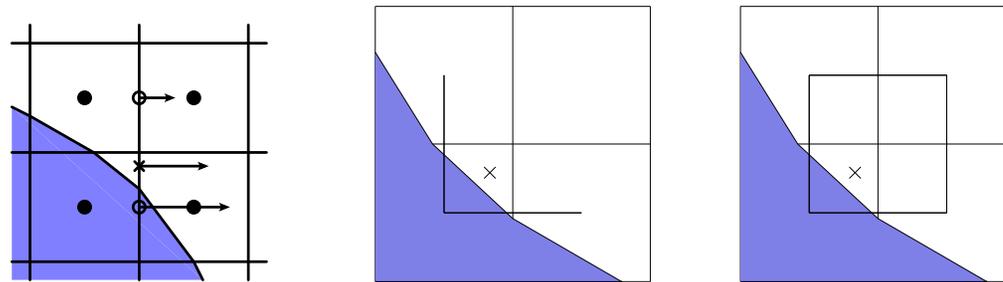


Smoothing of truncation error leads to a solution error that is $O(h^2)$ in max norm.

Extension to Three Dimensions

Our matrices aren't symmetric, nor are they M-matrices.

There are two obvious ways to extend the $O(h^2)$ flux calculation in 2D to 3D:

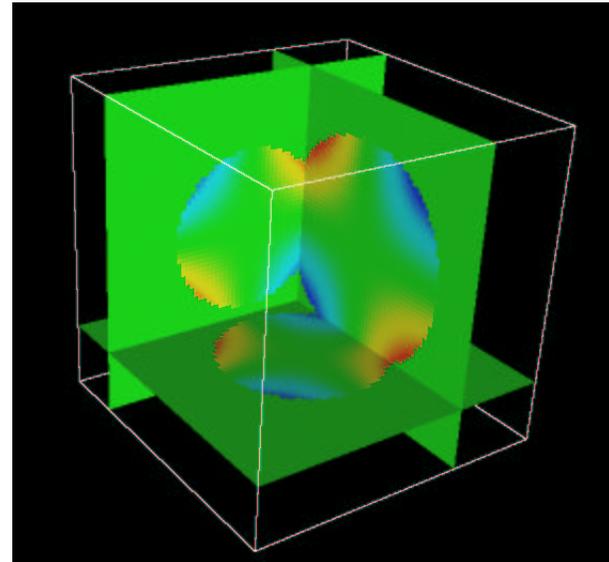
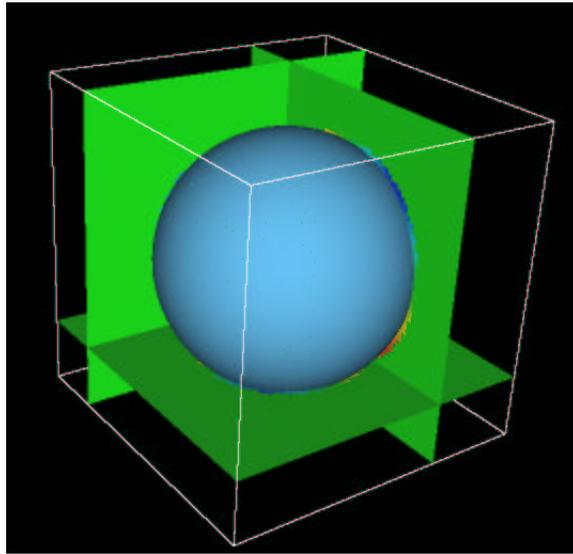


For intermittent configurations of adjacent small control volumes, linear interpolation is unstable (point Jacobi diverges), while bilinear interpolation appears to always be stable. Also, the inconsistent method coming from piecewise-constant interpolation is stable.

Unstable cases correspond to problems where small subproblems have eigenvalues of the wrong sign: the spectrum of PL^hP^t has elements in the right half-plane, where P is the projection onto a small set (2-8) of contiguous irregular control volumes.

Solution Error for Poisson's Equation in 3D

grid	$\ \epsilon\ _\infty$	p_∞	$\ \epsilon\ _2$	p_2	$\ \epsilon\ _1$	p_1
16^3	4.80×10^{-4}	—	5.17×10^{-5}	—	1.83×10^{-5}	—
32^3	1.06×10^{-4}	2.17	1.25×10^{-5}	2.05	4.41×10^{-6}	2.05
64^3	2.43×10^{-5}	2.13	3.07×10^{-6}	2.02	1.09×10^{-6}	2.02



Discretization of Hyperbolic Problems

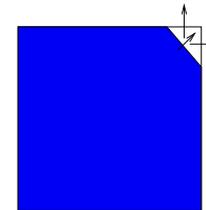
$$U^{n+1,h} = U^{n,h} - \Delta t D \cdot \vec{F}^{n+\frac{1}{2}}$$

Truncation error on irregular cells:

$$\tau \equiv \frac{U^{n+1,exact} - U^{n,exact}}{\Delta t} + D \cdot \vec{F}(U^{exact}) = O(h) + O(\Delta t^2) + O\left(\frac{h}{\kappa}\right)$$

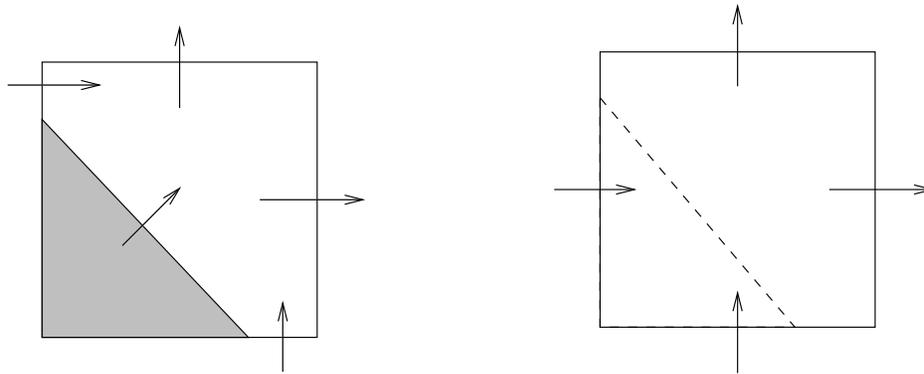
Want to use a time step given by the CFL for cells without the boundary.

$$\begin{aligned} U^{n+1} &= U^n - \Delta t D \cdot \vec{F} \\ &= U^n - \frac{\Delta t}{\kappa h} \left(\sum_{s \in faces} \alpha_s \vec{F}_s \cdot \vec{n}_s + \alpha_B \vec{F} \cdot \vec{n}_B \right) \end{aligned}$$



Flux Difference Redistribution

In irregular cells, we hybridize the conservative update $(D \cdot \vec{F})$ with a nonconservative, but stable scheme $(\nabla \cdot \vec{F})^{NC}$, and redistribute the nonconservative increment to nearby cells.



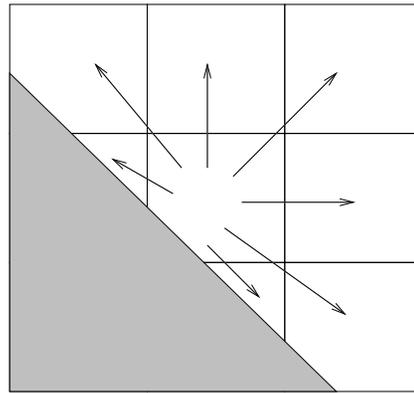
$$U^{n+1} = U^n - \Delta t(D \cdot \vec{F})^{NC} - w\Delta t((D \cdot \vec{F}) - (D \cdot \vec{F})^{NC})$$

The weight w is chosen so that, as $\kappa \rightarrow 1$, $w \rightarrow 1$, and $w = O(\kappa)$.

The amount of mass lost from each cell is

$$\delta M = -(1 - w)\kappa((\nabla \cdot \vec{F})^C - (\nabla \cdot \vec{F})^{NC}) = O(h)$$

We redistribute that mass to nearby cells in a volume-weighted way.



The truncation error for this method is $\tau = O(h)$ in cells sufficiently close to irregular cells, $\tau = O(h^2)$ otherwise.

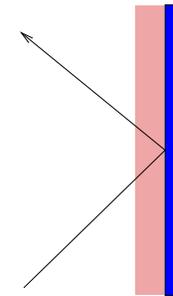
Modified Equation Analysis

$$U^h = U + \epsilon \approx U^{mod}$$

$$\frac{\partial U^{mod}}{\partial t} + \nabla \cdot \vec{F}(U^{mod}) = \tilde{\tau}$$

If the boundary is noncharacteristic, the large forcing on the boundary can only act for a short time:

$\frac{dU}{dt} = \tilde{\tau}$, but the characteristic path is in the region where $\tilde{\tau} = O(h)$ for only a time $O(h/\lambda)$, where λ is the characteristic speed. In that case,



$$U^h = U + O(h^2)$$

uniformly in x . If the boundary is characteristic, then we observe

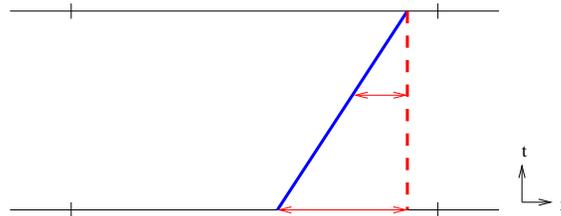
$$U^h = U + O(h) \text{ in } L^\infty \text{ norm}$$

$$U^h = U + O(h^2) \text{ in } L^1 \text{ norm}$$

Diffusion in a Time-Dependent Domain

$$\frac{\partial T}{\partial t} = \Delta T + f \text{ on } \Omega(t)$$

$$\frac{\partial T}{\partial n} = \dot{m} + sT \text{ on } \partial\Omega(t)$$



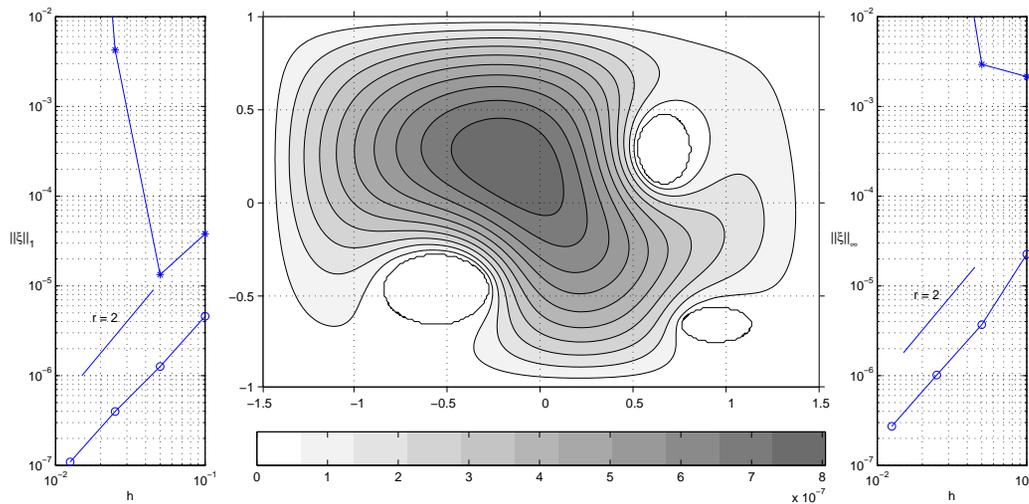
In order to use a second-order accurate implicit time discretization, it is necessary to convert the moving boundary problem into a sequence of fixed boundary problems.

- Move the boundary, updating cells that are uncovered by appropriate extrapolation.
- Solve the heat equation on a fixed domain for one time step, using extrapolated boundary conditions.

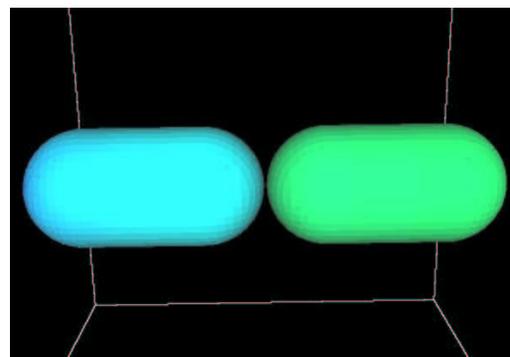
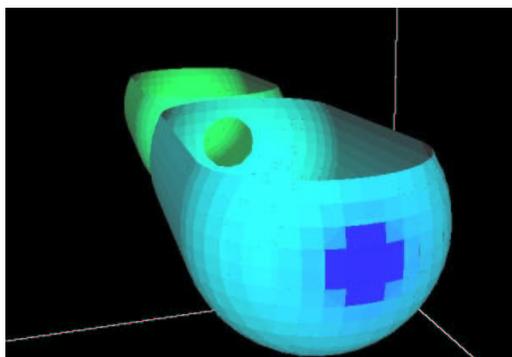
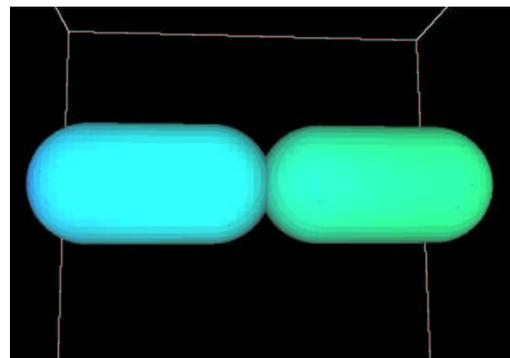
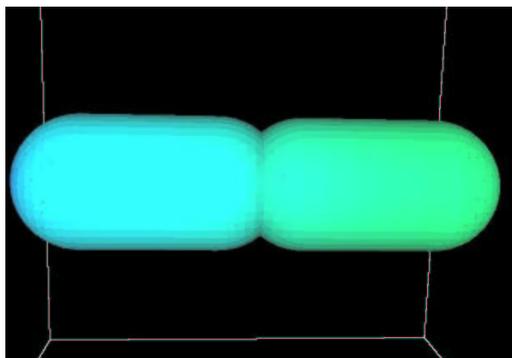
If we use Crank-Nicolson for the second step, the resulting method is unstable. To obtain a stable, second-order accurate method, must use an implicit Runge-Kutta method with better stability properties.

$$(I - r_1\Delta)(I - r_2\Delta)^{n+1} = (I + a\Delta)^n + \Delta t(I + r_4\Delta)f^{n+\frac{1}{2}}$$

$$r_1 + r_2 + a = \Delta t, r_1 + r_2 + r_4 = \frac{\Delta t}{2}, r_1 + r_2 > \frac{\Delta t}{2}$$



Moving Boundary Calculation in Three Dimensions



To treat more complex problems, we

- Decompose them into pieces, each one of which is well-understood, and between which the coupling is not too strong;
- Use numerical methods based on our understanding of the components, coupled together using predictor-corrector methods in time.

Example: Incompressible Navier-Stokes equations

$$\begin{aligned}\frac{\partial \vec{u}}{\partial t} + \vec{u} \cdot \nabla \vec{u} + \nabla p &= \nu \Delta \vec{u} \\ \nabla \cdot \vec{u} &= 0\end{aligned}$$

These equations can be splitting into three pieces:

$$\text{Hyperbolic: } \frac{\partial \vec{u}}{\partial t} + \vec{u} \cdot \nabla \vec{u} = 0$$

$$\text{Parabolic: } \frac{\partial \vec{u}}{\partial t} = \nu \Delta \vec{u}$$

$$\text{Elliptic: } \Delta p = \nabla \cdot (-\vec{u} \cdot \nabla \vec{u} + \nu \Delta \vec{u})$$

Problems Arising in Decomposition into Classical Components

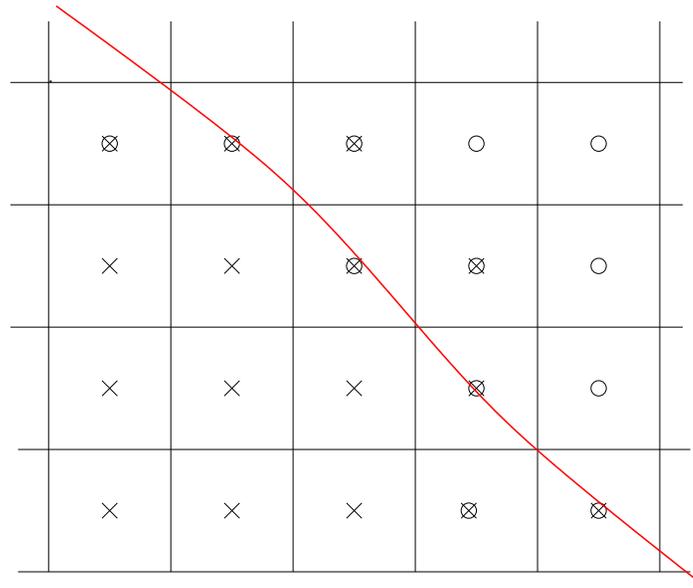
Using asymptotics to eliminate fast scales, or split slow and fast scales.

- Low Mach number asymptotics to eliminate acoustic scales: incompressible flow, low-Mach-number combustion, anelastic models for geophysical flows (Rehm and Baum, 1978; Majda and Sethian, 1985; Lai, Bell, Colella, 1993).
- Allspeed methods - splitting dynamics into vortical and compressive components (Colella and Pao, 1999).
- Methods for splitting the fast dielectric relaxation dynamics in charged-fluid models of “almost” quasineutral plasmas (Vitello and Graves, 1997; Colella, Dorr, and Wake, 1999).

All of these approaches lead to the introduction of redundant equations or constraints: $p = const.$, $\dot{q}_{net} = \dots$. The presence of such constraints complicate the formulation of time-discretization methods.

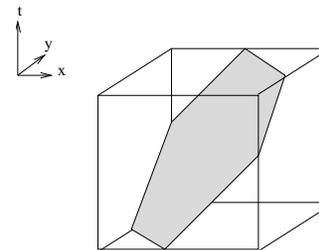
Hyperbolic PDEs containing gauge constraints, such as ideal MHD ($\nabla \cdot \vec{B} = 0$) or solid mechanics, are well-posed only if the constraint is satisfied. Truncation error may cause the constraint to be violated (Miller and Colella, 2001; Crockett, et. al., to appear).

Cartesian Grid Discretization of Free Boundary Problems



- Solution is double-valued on all cells intersecting the free boundary.
- Finite-volume discretization of conservation laws on each control volume on either side of the front.
- Motion of the front and discretization in the interior are coupled via the jump relations:

$$\kappa^{n+1}U^{n+1} = \kappa^n U^n + \{ \text{sum of fluxes} \}$$

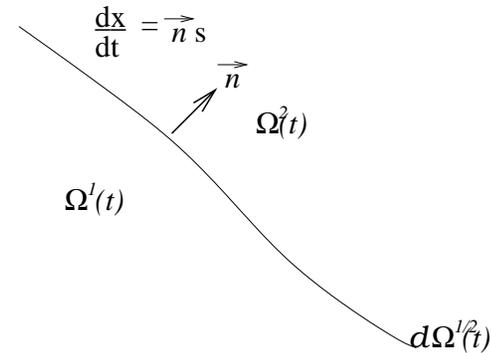


Hyperbolic Free Boundary Problems

$$\frac{\partial U^q}{\partial t} + \nabla \cdot \vec{F}^q = 0 \text{ on } \Omega^q(t), \quad q = 1, 2$$

$$[\vec{F} \cdot \vec{n} - sU] = g \text{ on } \partial\Omega^{1/2}(t)$$

$$[f] \equiv f^2 - f^1$$



- Discrete geometric quantities are a function of time, e.g., $\kappa = \kappa(t)$.
- Divergence theorem is applied in space-time to obtain discrete evolution equation:

$$0 = \int \frac{\partial U}{\partial t} + \nabla \cdot \vec{F} dxdt =$$

$$\bar{\kappa}^{n+1} \bar{U}^{n+1} - \bar{\kappa}^n \bar{U}^n + \frac{\Delta t}{h} \left(\sum_{s \in \text{faces}} \bar{\alpha}_s \vec{F}_s \cdot \vec{n}_s + \bar{\alpha}_B (\vec{F} \cdot \vec{n}_B - sU) \right)$$

Changes to Fixed-Boundary Algorithm

- Riemann problem used to compute fluxes, speed of the front.
- Small-cell stability: mass increments are redistributed along characteristics in the direction normal to front.

$$\delta M = \delta M^+ + \delta M^-$$

$$\delta M^+ = \sum_{\lambda_k \geq 0} (l_k \cdot \delta M) r_k$$

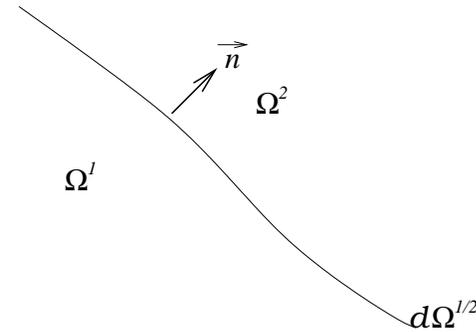
δM^+ remains on the same side of the front as it was generated on, while δM^- is redistributed across the front.

- Accuracy: for genuinely nonlinear waves, free boundary is noncharacteristic, so solution error is one order smaller than truncation error in max norm.

Elliptic Free Boundary Problems

$$\beta \Delta \phi^q = \rho^q \text{ on } \Omega^q, \quad q = 1, 2$$

$$\left[\beta \frac{\partial \phi}{\partial n} \right] = g_N, \quad [\phi] = g_D \text{ on } \partial \Omega^{1/2}$$

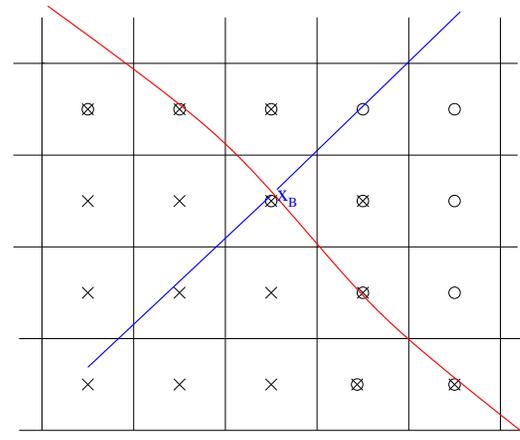


Given the values at the cell centers, the algorithm for the fixed boundary can be used to evaluate the operator, provided that one can find the values for ϕ_B^q . The jump relations lead to a pair of linear equations for ϕ_B^q :

$$\phi_B^1 - \phi_B^2 = g_D(\vec{x}_B)$$

$$\beta^1 \frac{d\Phi^1}{dr} - \beta^2 \frac{d\Phi^2}{dr} = g_N(\vec{x}_B)$$

Where $\Phi^q(r)$ are the interpolating polynomials along the normal directions from \vec{x}_B .



Free Boundary Problems for Diffusion

$$\frac{\partial T^\alpha}{\partial t} = D^\alpha \Delta T^\alpha + f^\alpha \text{ on } \Omega^\alpha(t), \alpha = 1, 2$$

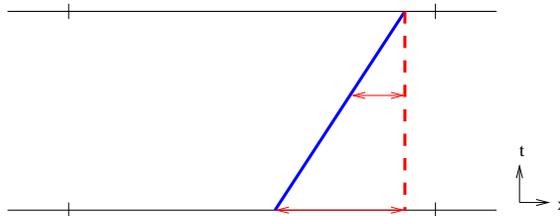
$$\left[D \frac{\partial T}{\partial n} \right] = g_N, [T] = g_D \text{ on } \partial\Omega^{1/2}(t)$$

s is prescribed (*not* the Stefan problem).

As before, we convert a moving boundary problem into a sequence of problems on fixed boundaries.

$$\left[D \frac{\partial T}{\partial n} \right] = g_N + \vec{\delta} \cdot \left[D \nabla \frac{\partial T}{\partial n} \right]$$

$$[T] = g_D + \vec{\delta} \cdot [\nabla T] + \frac{1}{2} \vec{\delta} \cdot [\nabla \nabla T] \cdot \vec{\delta}$$



Future Work and Open Questions

- Adaptive mesh refinement.
- Software infrastructure.
- Decomposition into classical components: phase change boundaries, surface tension.
- Consistent discretization methods for free-boundary case.
- Other applications: magnetic fusion, combustion, cell modeling, bio-MEMS.